## Amendments to the Claims

Please amend the claims as follows:

1. (Currently Amended): A method for providing predictive maintenance of a device, comprising the steps of:

modeling as a time series  $x_n$  of a discretely sampled signal representative of occurrences of a defined event in the operation of said device, said time series  $x_n$  being modeled as two-state first order Markov processes with associated transition probabilities p(i|j), wherein state 1 applies when the number of said occurrences exceeds a certain threshold T, and state 0 applies when the number of said occurrences falls below said certain threshold T, being represented as:

$$S_n = \begin{cases} 0 & \text{if} \quad x_n \le T \\ 1 & \text{if} \quad x_n > T \end{cases}$$

wherein said transition probability p(i|j) is the switching probability from state j to state i, that is, the probability that  $S_n = i$  given that  $S_{n-1} = j$ , being a total of 4 transition probabilities;

computing said four transition probabilities the last N states  $S_n$ , where N is a predetermined number;

conducting a supervised training session utilizing a set of J devices, which have failed due to known causes and considering the two independent probabilities p(1|1) and p(1|0), said training session comprising:

computing the two-dimensional feature vectors  $f_i = \{p(1|1), p(1|0)\}_i$  for the initial M windows of N scans,

computing the two-dimensional feature vectors  $f_f = \{p(1|1), p(1|0)\}_f$  for the final N number of scans,

plotting a scatter-diagram of all 2D feature vectors  $(f_i)_n$  and  $(f_f)_n$ , (n = 1...J), and

deriving <u>from the scatter-diagram</u> a pattern classifier by estimating the optimal linear discriminant which separates the two foregoing sets of vectors; and

applying said classifier to monitor the persistence of occurrences of said defined event in the operation of said device.

2. (Currently Amended): A method for providing predictive maintenance of a device as recited in claim 1, including the steps of:

updating said transition probabilities at each scan-are-updated; and constructing the feature vector  $f = \{p(1|1), p(1|0)\}$ -constructed.

3. (Currently Amended): A method for providing predictive maintenance of a device as recited in claim 2, including the step of:

providing a warning of imminent failure of said device if f falls into a region of said classifier <del>corresponding</del> indicating such failure prediction.

4. (Currently Amended): A method for providing predictive maintenance of an X-ray tube, comprising the steps of:

modeling as a time series  $x_n$  of a discretely sampled signal representative of occurrences of arcing in the operation of said tube, said time series  $x_n$  being modeled as two-state first order Markov processes with associated transition probabilities p(i|j), wherein state 1 applies when the number of said occurrences exceeds a certain threshold T, and state 0 applies when the number of said occurrences falls below said certain threshold T, being represented as:

$$S_n = \begin{cases} 0 & \text{if} \quad x_n \le T \\ 1 & \text{if} \quad x_n > T \end{cases}$$

wherein said transition probability p(i|j) is the switching probability from state j to state i, that is, the probability that  $S_n = i$  given that  $S_{n-1} = j$ , being a total of 4 transition probabilities;

computing said four transition probabilities the last N states  $S_n$ , where N is a predetermined number;

conducting a supervised training session utilizing a set of J X-ray tubes, which have failed due to known causes and considering the two independent probabilities p(1|1) and p(1|0), said training session comprising:

computing the two-dimensional feature vectors  $f_i = \{p(1|1), p(1|0)\}_i$  for the initial M windows of N scans,

computing the two-dimensional feature vectors  $f_f = \{p(1|1), p(1|0)\}_f$  for the final N number of scans,

plotting a scatter-diagram of all 2D feature vectors  $(f_i)_n$  and  $(f_f)_n$ , (n = 1...J), and

deriving a pattern classifier by estimating the optimal linear discriminant which separates the two foregoing sets of vectors; and

applying said classifier to monitor the persistence of occurrences of said arcing in the operation of said X-ray tube.

5. (Original): A method for providing predictive maintenance of an X-ray tube as recited in claim 4, including the steps of:

updating said transition probabilities at each scan are updated; and constructing the feature vector  $f = \{p(1|1), p(1|0)\}$  constructed.

6. (Currently Amended): A method for providing predictive maintenance of an X-ray tube as recited in claim A5, including the step of:

providing a warning of imminent failure of said X-ray tube if f falls into a region of said classifier corresponding indicating such failure prediction.

7. (Currently Amended): A method for providing predictive maintenance of a device eomprises comprising the steps of modeling as a time series of a discretely sampled signal representative of occurrences of a defined event in the operation of said device, said time series being modeled as two-state first order Markov processes with associated transition probabilities, wherein one state applies when the number of said occurrences exceeds a certain threshold, and the other state applies when the number of said occurrences falls below said certain threshold; computing said four transition probabilities the last N states S<sub>n</sub>, where N is a predetermined number, conducting a supervised training session utilizing a set of J devices, which have failed due to known causes and considering the two independent probabilities and, said training session comprising computing the two-

dimensional feature vectors for the initial M windows of N scans, computing the twodimensional feature vectors for the final N number of scans, plotting a scatter-diagram of all 2D feature vectors, and deriving a pattern classifier by estimating the optimal linear discriminant which separates the two foregoing sets of vectors; and applying said classifier to monitor the persistence of occurrences of said defined event in the operation of said device.

8. (Currently Amended): Apparatus for providing predictive maintenance of a device, comprising:

means for modeling as a time series  $x_n$  of a discretely sampled signal representative of occurrences of a defined event in the operation of said device, said time series  $x_n$  being modeled as two-state first order Markov processes with associated transition probabilities p(i|j), wherein state 1 applies when the number of said occurrences exceeds a certain threshold T, and state 0 applies when the number of said occurrences falls below said certain threshold T, being represented as:

$$S_n = \begin{cases} 0 & \text{if} \quad x_n \le T \\ 1 & \text{if} \quad x_n > T \end{cases}$$

wherein said transition probability p(i|j) is the switching probability from state j to state i, that is, the probability that  $S_n = i$  given that  $S_{n-1} = j$ , being a total of 4 transition probabilities;

means for computing said four transition probabilities the last N states  $S_n$ , where N is a predetermined number;

means for conducting a supervised training session utilizing a set of J devices, which have failed due to known causes and considering the two independent probabilities p(1|1) and p(1|0), said means for conducting a supervised training session comprising means for:

computing the two-dimensional feature vectors  $f_i = \{p(1|1), p(1|0)\}_i$  for the initial M windows of N scans,

computing the two-dimensional feature vectors  $f_f = \{p(1|1), p(1|0)\}_f$  for the final N number of scans,

plotting a scatter-diagram of all 2D feature vectors  $(f_i)_n$  and  $(f_f)_n$ ,  $(n=1\dots J)$ , and

deriving a pattern classifier by estimating the optimal linear discriminant which separates the two foregoing sets of vectors; and

means for applying said classifier to monitor the persistence of occurrences of said defined event in the operation of said device.